

✓ Color

We look into two principal uses of color scales:

- to distinguish unordered qualitative attributes; and
- to represent quantitative data values.

A *color scale*, also called a *colormap*, is a mapping from the range of an attribute's data values to a color range. In the case of continuous color scales, indexes into the color scale are interpolated.

Plotly references on color: [continuous color scales](#), [discrete color scales](#), and [built-in color scales](#).

```
1 from google.colab import drive
2 drive.mount('/content/drive')
3
4 # datadir = 'data/'
5 # imagesdir = 'images/'
6
7
8 from IPython.display import Image, display
9
10 datadir = '/content/drive/My Drive/Courses/672/Notebooks/Notebooks2023/data/'
11 imagesdir = '/content/drive/My Drive/Courses/672/Notebooks/Notebooks2023/images/'
12
13 def display_images(images, dir=imagesdir):
14     for image in images:
15         display(Image(dir + image))
```

 Mounted at /content/drive

```
1 import plotly.express as px
2 import plotly.graph_objects as go
3 import numpy as np
4
5 rendering = None
6
7 def show(fig):
8     fig.show(rendering=rendering)
```

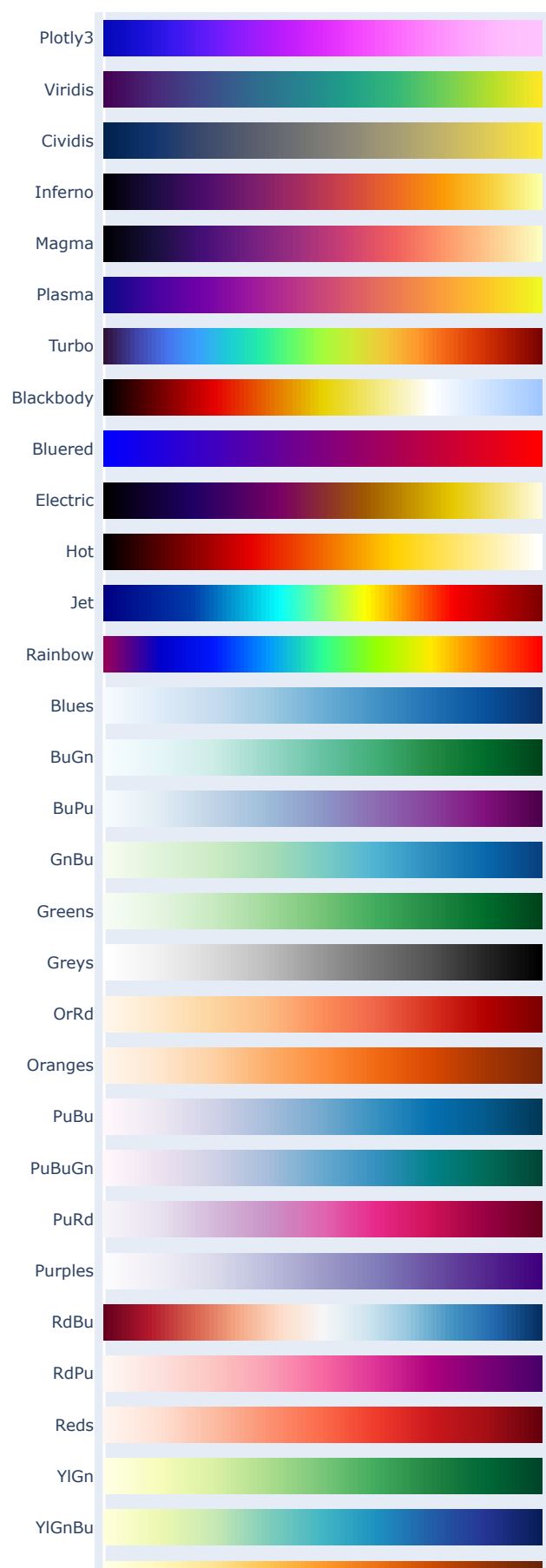
✓ Continuous color scales

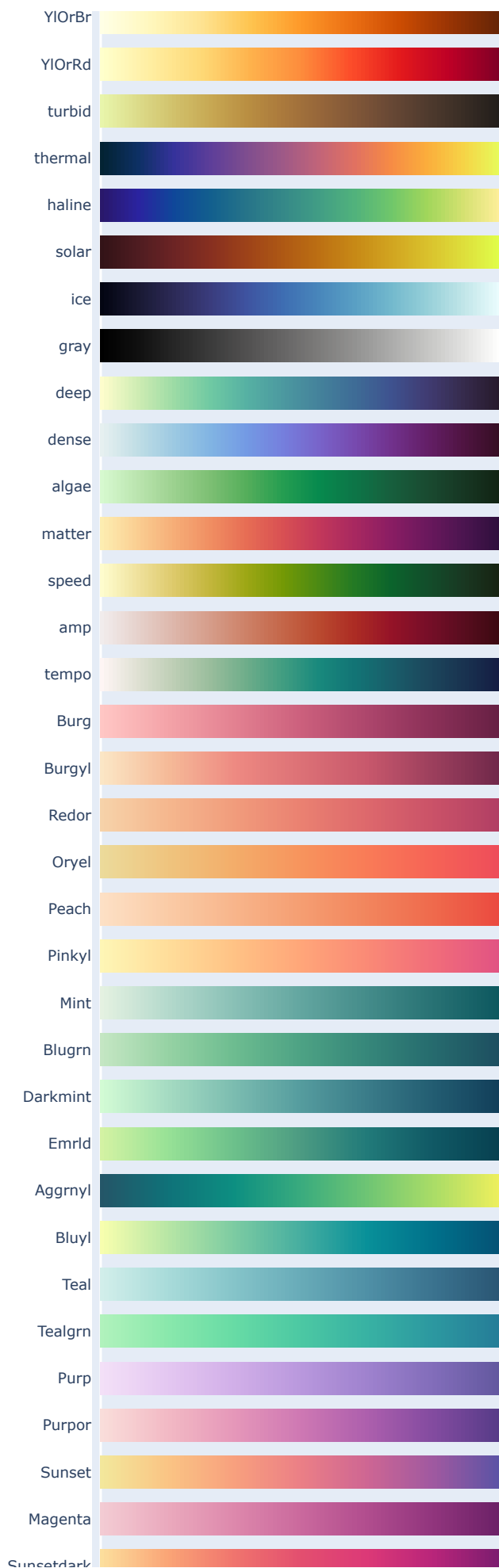
Continuous color scales are used to represent continuous numeric data values. Here we view the built-in continuous color scales.

```
1 fig = px.colors.sequential.swatches_continuous()
2 show(fig)
```



plotly.colors.sequential



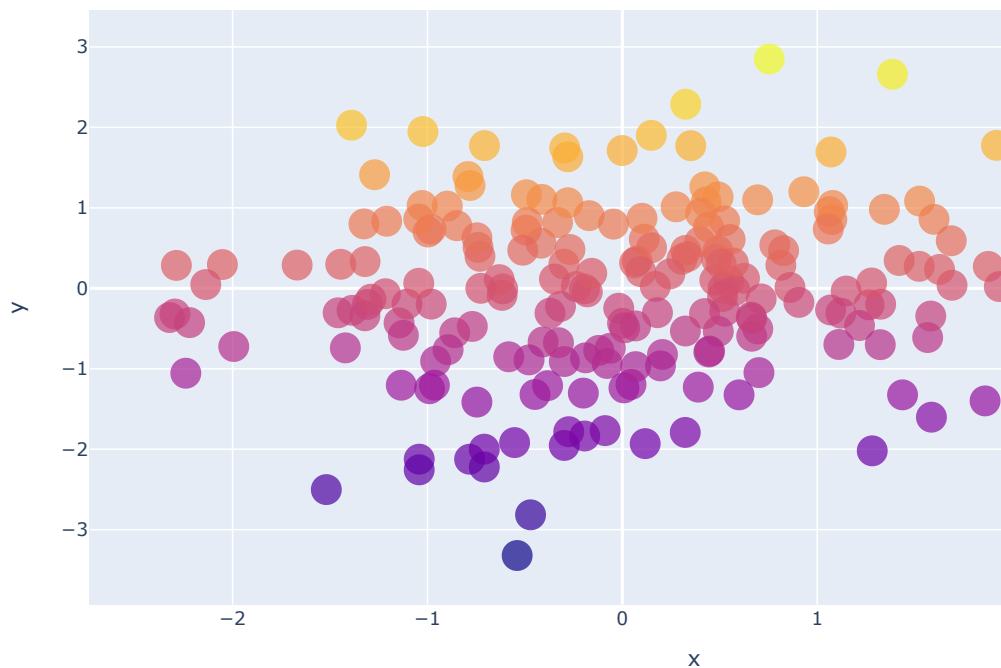




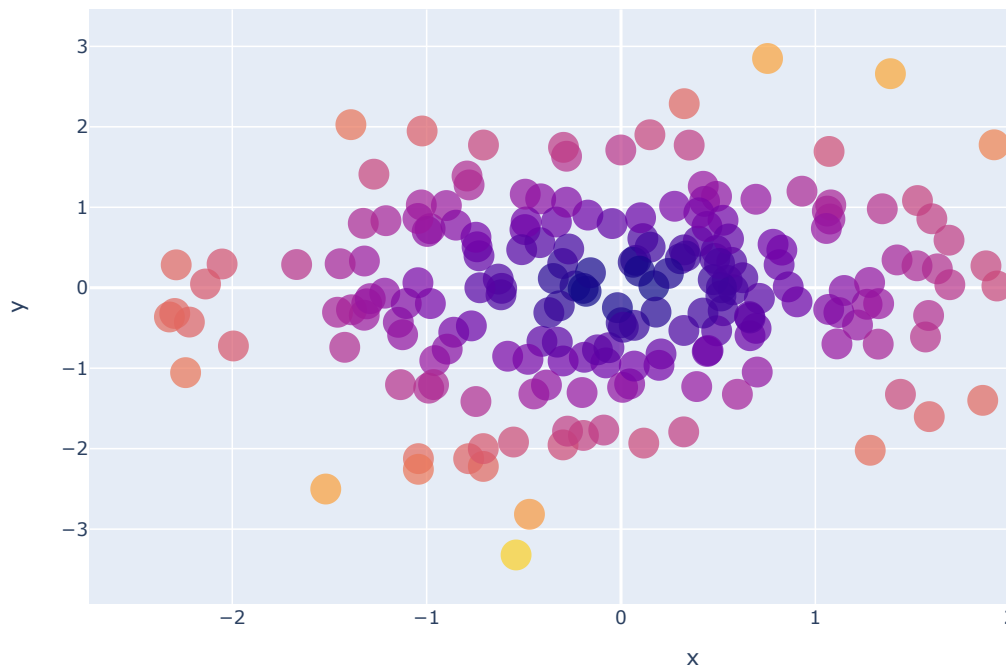
We generate some normally-distributed points in the plane. When we generate a plot, the `color` argument supplies a value for each point corresponding to its color under the current color scale which defaults to *Plotly*.

```
1 nbr_points = 200
2 x, y = np.random.normal(0, 1, nbr_points), np.random.normal(0, 1, nbr_points)

1 size = 20
2 fig = px.scatter(x=x, y=y, opacity=0.7, color=y)
3 fig.update_traces(marker_size=size)
4 show(fig)
```

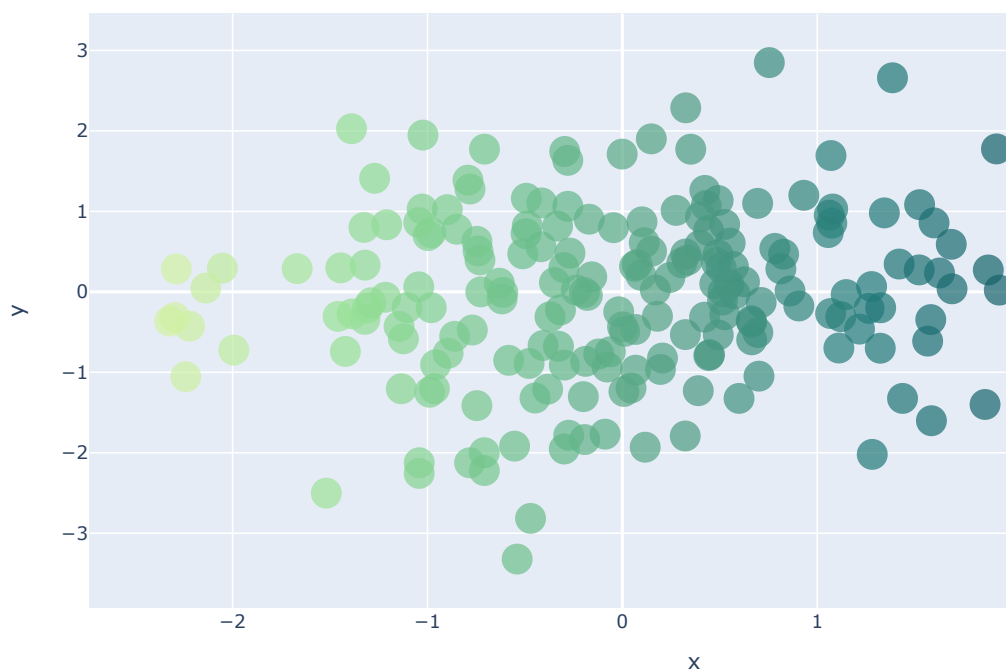


```
1 def norm(x, y):
2     return np.sqrt(x**2 + y**2)
3 fig = px.scatter(x=x, y=y, opacity=0.7, color=norm(x, y))
4 fig.update_traces(marker_size=size)
5 show(fig)
```

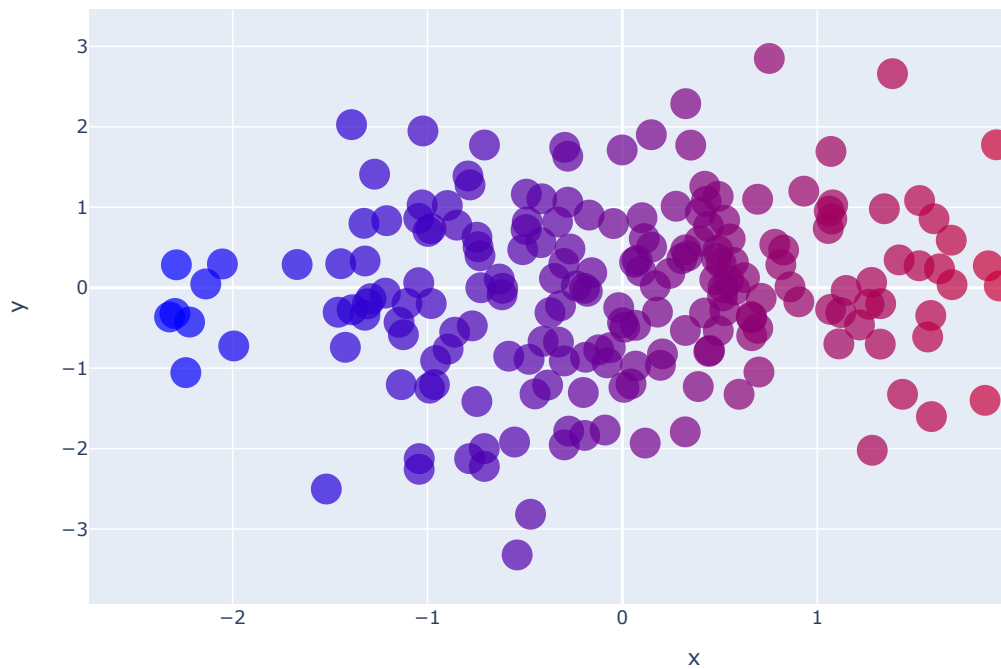


For many plotting functions, we use the `color_continuous_scale` argument to select a different color scale.

```
1 fig = px.scatter(x=x, y=y, opacity=0.7, color=x, color_continuous_scale=px.colors.sequential.Emrld)
2 # fig = px.scatter(x=x, y=y, opacity=0.7, color=x, color_continuous_scale='Rainbow')
3 fig.update_traces(marker_size=size)
4 show(fig)
5
```



```
1 fig = px.scatter(x=x, y=y, opacity=0.7, color=x, color_continuous_scale='Bluered')
2 fig.update_traces(marker_size=size)
3 show(fig)
```

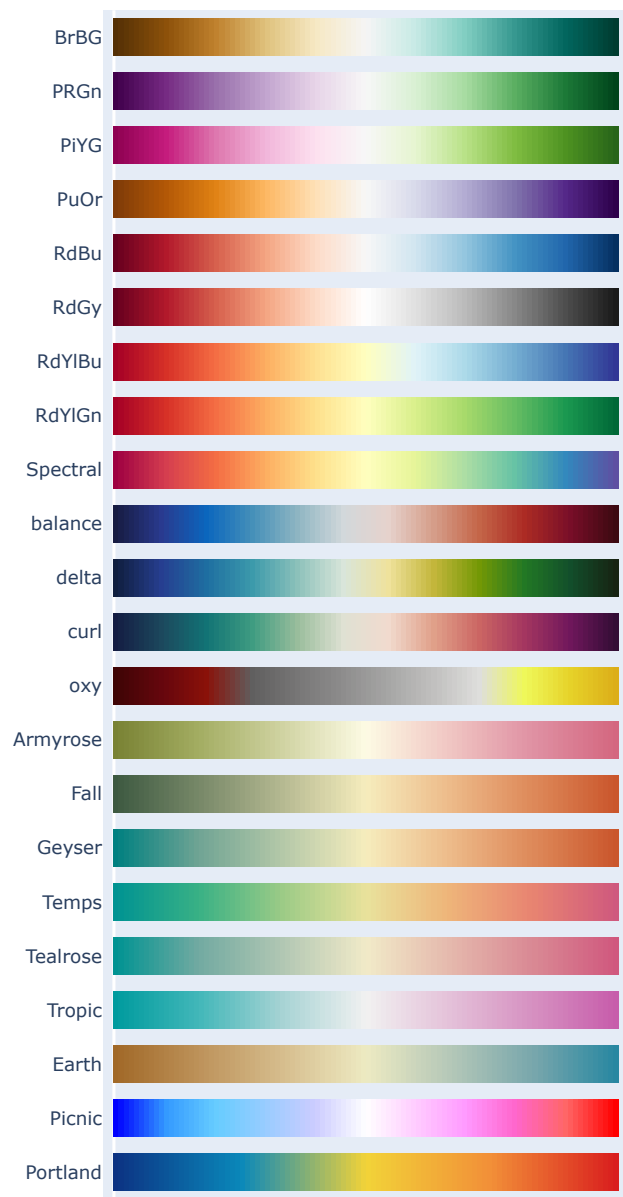


Continuous color scales are used when the data values contain no special *zero* value. We use a *diverging* color scale when we want to distinguish between values less than zero and values greater than zero.

```
1 show(px.colors.diverging.swatches_continuous())
```



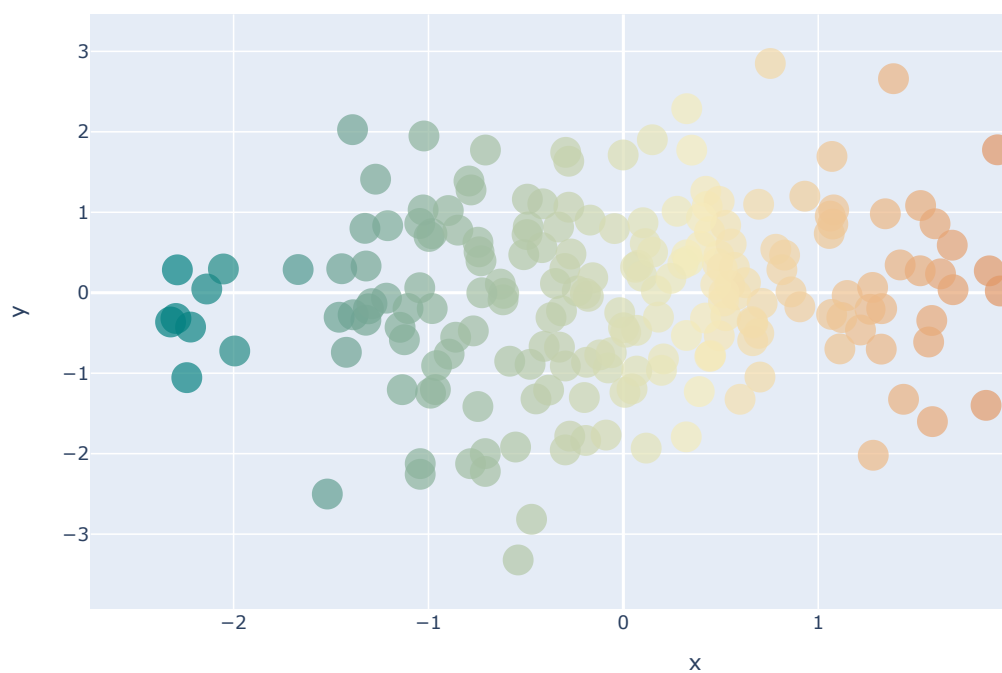
plotly.colors.diverging



```

1 fig = px.scatter(x=x, y=y, opacity=0.7, color=x, color_continuous_scale='Geyser')
2 fig.update_traces(marker_size=size)
3 show(fig)

```

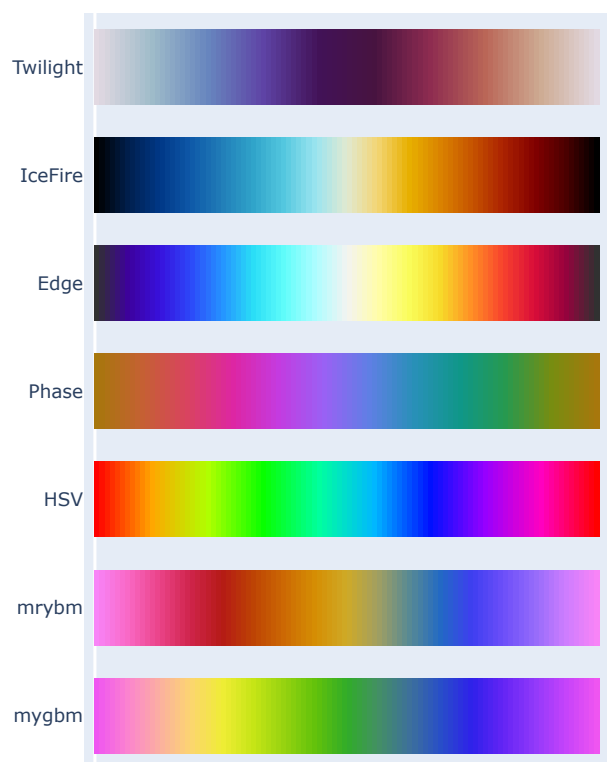


Cyclic color scales are used for data that is cyclic.

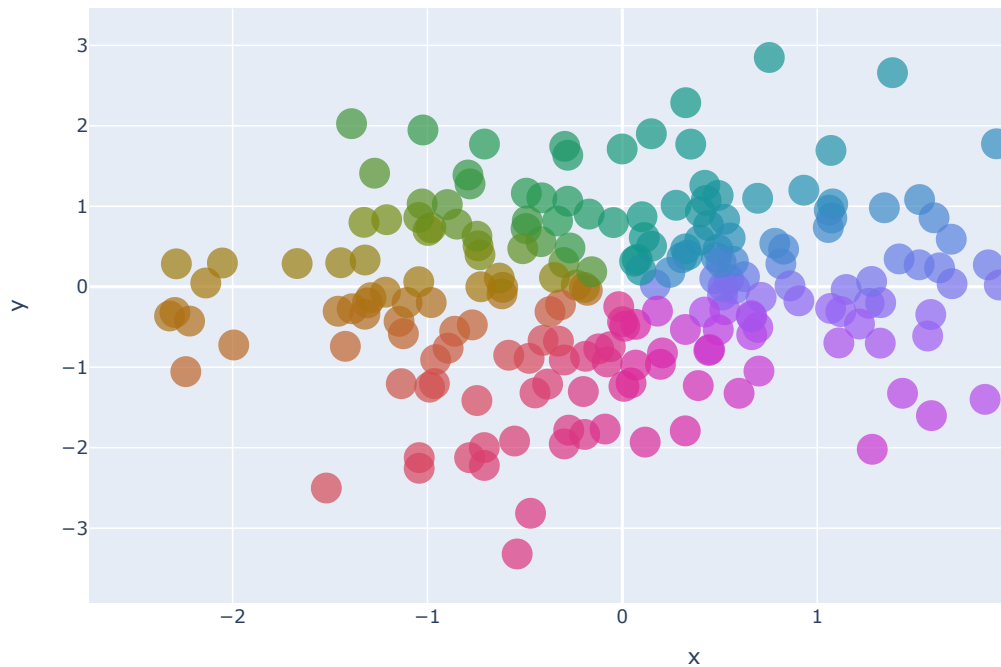
```
1 show(px.colors.cyclical.swatches_continuous())
```



plotly.colors.cyclical




```
1 def angle(x, y):  
2     return np.arctan2(x, y)  
3  
4 fig = px.scatter(x=x, y=y, opacity=0.7, color=angle(y,x), color_continuous_scale=px.colors.cyclical.Phase)  
5 fig.update_traces(marker_size=size)  
6 show(fig)
```



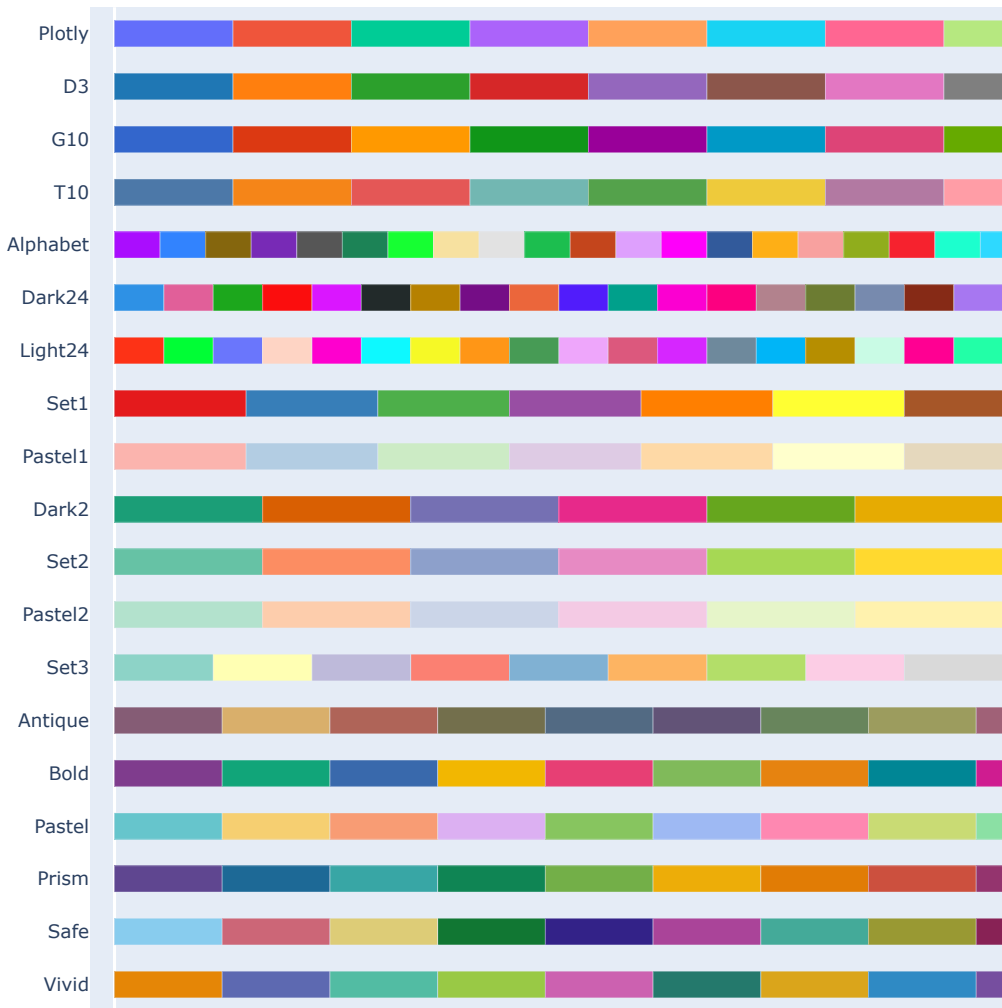
✓ Discrete color scales

Discrete color scales are used to represent categorical or discrete data values.

```
1 show(px.colors.qualitative.swatches())
```



plotly.colors.qualitative

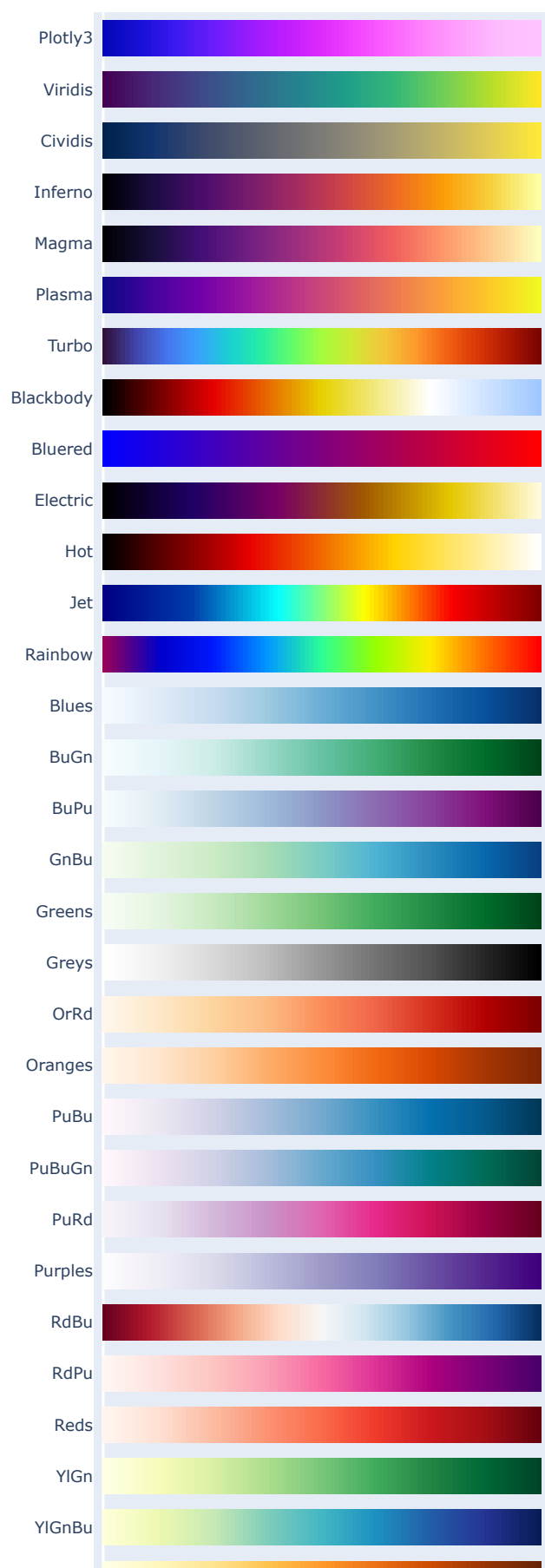


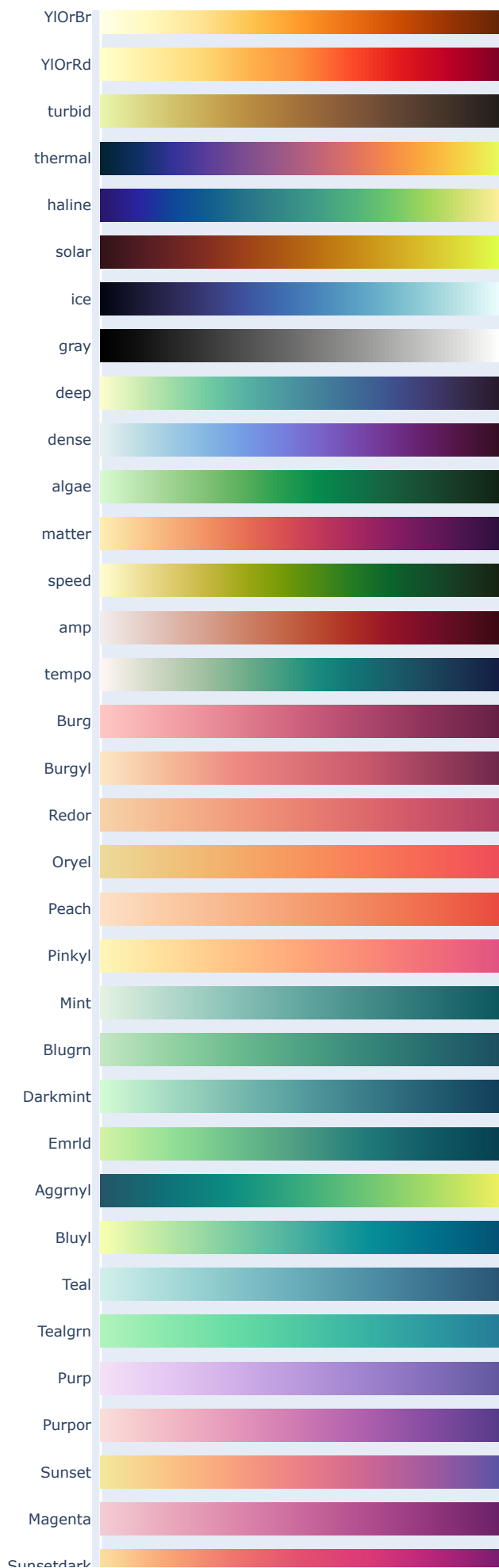
We also have discrete versions of the continuous colormaps we've already seen.

```
1 show(px.colors.sequential.swatches_continuous())
2 show(px.colors.sequential.swatches())
```



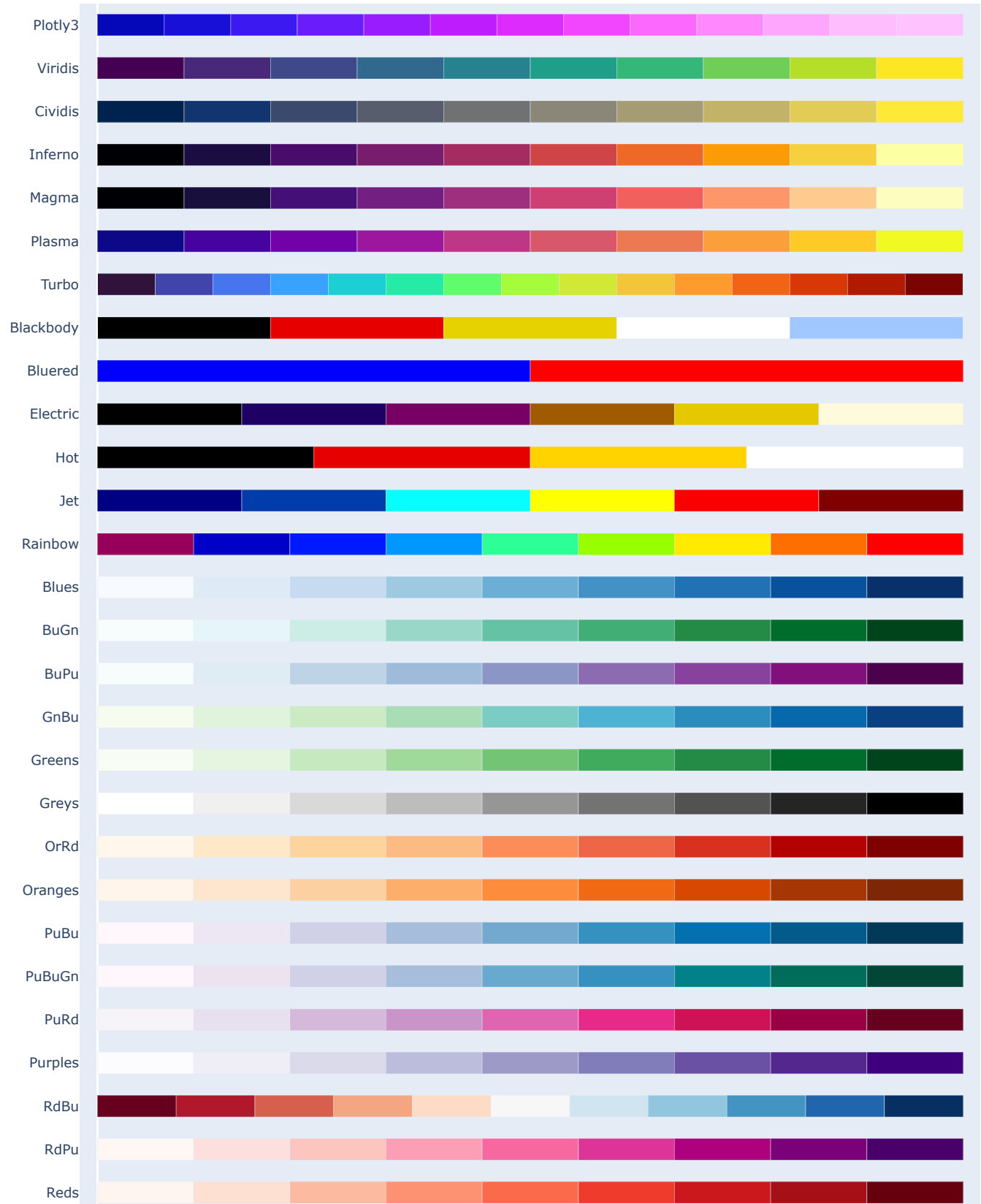
plotly.colors.sequential

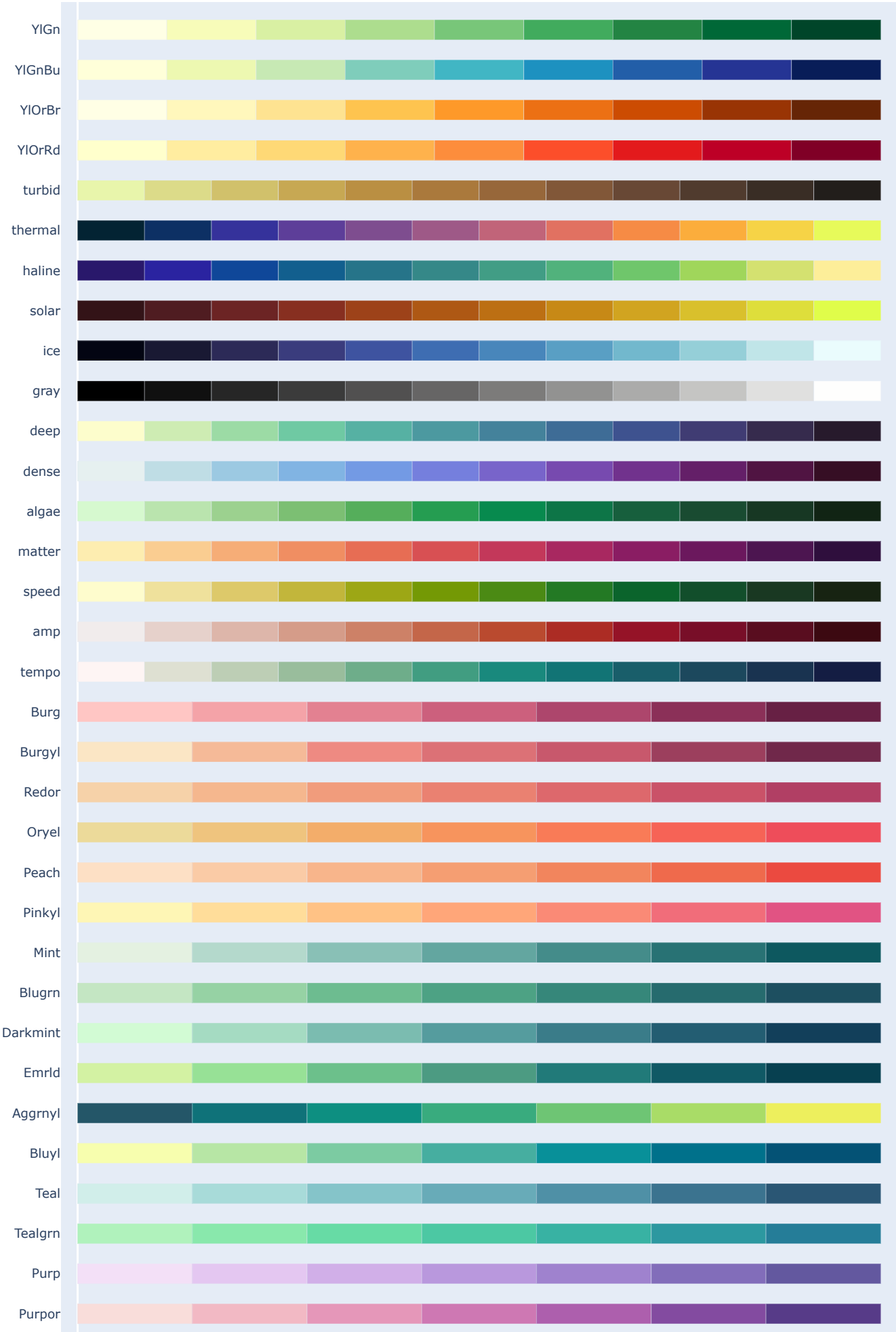


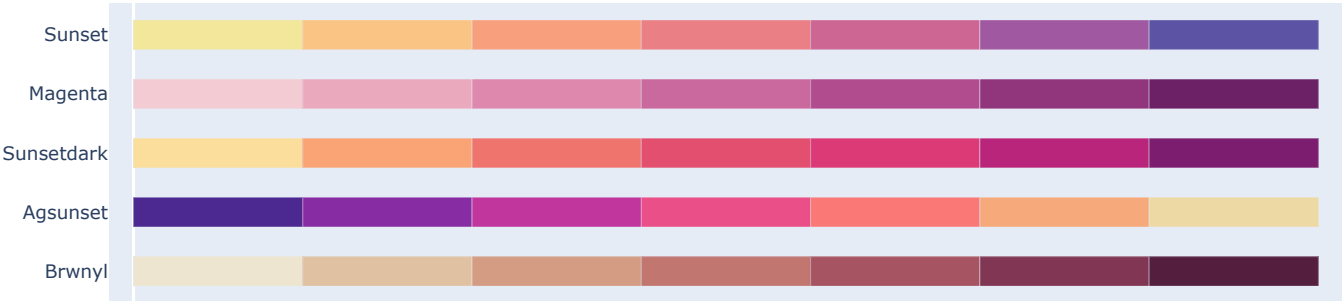




plotly.colors.sequential







```
1 df = px.data.iris() # iris is a pandas DataFrame
2 df
```

	sepal_length	sepal_width	petal_length	petal_width	species	species_id
0	5.1	3.5	1.4	0.2	setosa	1
1	4.9	3.0	1.4	0.2	setosa	1
2	4.7	3.2	1.3	0.2	setosa	1
3	4.6	3.1	1.5	0.2	setosa	1
4	5.0	3.6	1.4	0.2	setosa	1
...
145	6.7	3.0	5.2	2.3	virginica	3
146	6.3	2.5	5.0	1.9	virginica	3
147	6.5	3.0	5.2	2.0	virginica	3
148	6.2	3.4	5.4	2.3	virginica	3
149	5.9	3.0	5.1	1.8	virginica	3

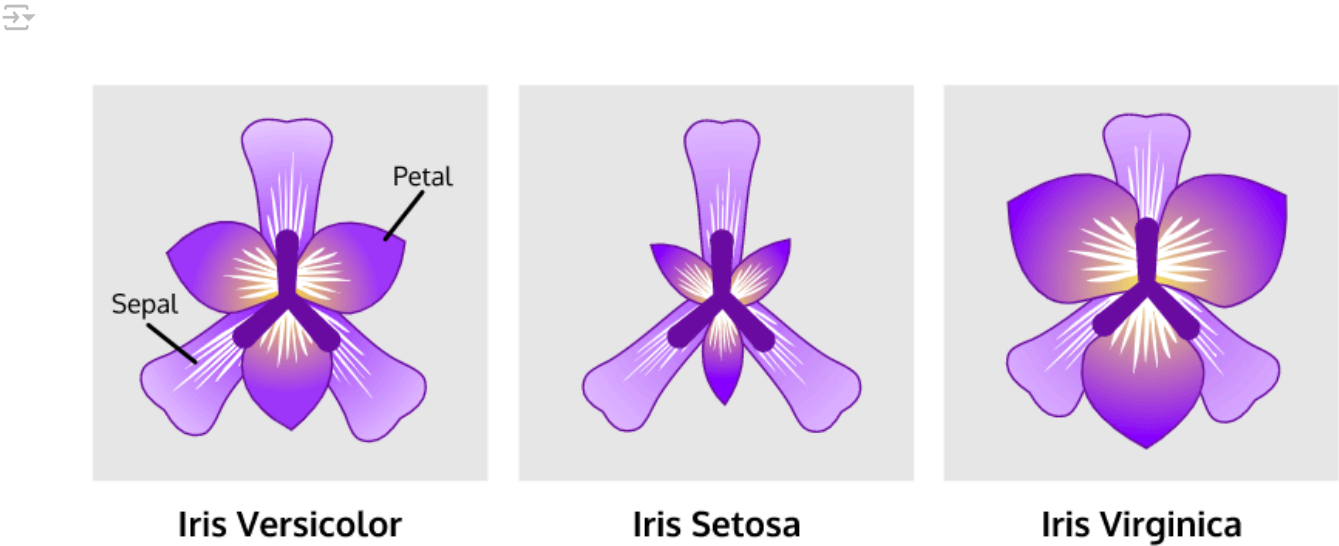
150 rows x 6 columns

Next steps:

Generate code with df

☒ View recommended plots

```
1 display_images(['iris.png'])
```



```

1 # use color to distinguish species (discrete)
2 fig = px.scatter(df, x="sepal_width", y="sepal_length", color='species', color_discrete_sequence=px.colors.
3 # some discrete colormaps: color_discrete_sequence=px.colors.qualitative.*: Bold, Alphabet
4 # fig = px.scatter(df, x="sepal_width", y="sepal_length", color='species', color_discrete_sequence=px.color
5 # fig = px.scatter(df, x="sepal_width", y="sepal_length", color='species')
6 # some discrete colormaps: color_discrete_sequence=px.colors.sequential.*: RdBu, Tealgrn, BuPu, PuBu
7 fig.update_traces(marker_size=10)
8 show(fig)

```

